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## Introduction

## Numerical Methods

Eulers Method:

 $y_{n+1} = y_n + hf(t_n, y_n)$ 

## Convergence of Euler's Method

$$\begin{split} \left| \frac{\partial f}{\partial y} \right| &\leq L \\ L &\leq \max_{(t,y) \in R} \left| \frac{\partial f}{\partial y} \right| \\ \left| y''(t) \right| &\leq M, \quad a \leq t \leq b \\ D &= e^{(b-a)L} \frac{M}{2L} \end{split}$$

and with  $e_n = y_n - y(t_n)$  we have the global error is bounded by Dh in magnitude:  $|e_n| \leq Dh$ , for n = 0, 1, ..., N.

## The Flow Map

**Flow Map:** fixing  $t_0$  and h we may consider the map from  $y_0$  to  $y(t_0 + h; t_o, y_0)$ . This is the flow map, written:

$$y(t_0 + h; t_0, y_0) = \Phi_{t_0, h}(y_0)$$

Which is actually a family of maps. **Flow Map Approximation:** can be viewing of the form:

$$\hat{\Phi}_{t,h}(y) = y + hf(t,y)$$

Methods which approximate the solution through iteration of an approximate flow map are called one-step methods.

## **Taylor Series Methods**

$$y' = f(t, y)$$
$$y'' = f_y y' + f_t = f_y f + f_t$$

Thus

$$\Phi_{t,h}(y) = y + hf(t,y) + \frac{1}{2}h^2 \left(f_y(t,y)f(t,y) + f_t(t,y)\right) + \frac{1}{6}y'''h^3 + \dots$$

Convergence of One-Step Methods For a method to have order p = 3 it must satisfy p = 2 and:

## Polynomial Interpolation

The Lagrange interpolating polynomials  $\ell_i, i = 1, \ldots, s$  for a set of abscissae are defined by:

$$\ell_i(x) = \prod_{\substack{j=1\\j\neq i}}^s \frac{x - c_j}{c_i - c_j}$$

Defining the weights:

$$b_i = \int_0^1 \ell_i(x) \mathrm{d}x$$

The quadrature formula becomes:

$$\int_0^1 g(x) \mathrm{d}x \approx \int_0^1 P(x) \mathrm{d}x = \sum_{i=1}^s b_i g\left(c_i\right)$$

# Runge-Kutta Methods

$$Y_{i} = y_{n} + h \sum_{j=1}^{s} a_{ij} f(Y_{j}), \quad i = 1, \dots, s$$
  
 $y_{n+1} = y_{n} + h \sum_{i=1}^{s} b_{i} f(Y_{i})$ 

Here, s is the number of stages of the RK method,  $b_i$  are the weights and  $a_{ij}$  are the internal coefficients. Butcher Tables:

## Order Conditions

For a method to have order p = 1 we need:

$$\sum_{i=1}^{s} b_i = 1$$

For a method to have order p = 2, it must satisfy order p = 1 and:

$$\sum_{i=1}^{s} b_i c_i = \frac{1}{2}$$

 $\sum_{i=1}^{s} b_i c_i^2 = \frac{1}{3}, \quad \sum_{i=1}^{s} \sum_{j=1}^{s} b_i a_{ij} c_j = \frac{1}{6}$ 

## Equilibrium points

**Equilibrium Points:** is a point of  $\frac{dy}{dt} = f(y)$  for which  $f(y^*) = 0$ .

An equilibrium point of the ODE corresponds to a fixed point of the flow map. A point  $y^* \in \mathbb{R}^d$  such that

$$\phi_t(y^*) = y^*, \forall t > 0$$

#### Asymptotically Stable: if

- $y^*$  is stable
- Solutions started sufficiently near to  $y^*$  tend to  $y^*$  as  $t \to \infty$

**Theorem 4.3.2:** Suppose  $f(y) = \frac{dy}{dt}$  is  $C^2$  and has an equilibrium point  $y^*$ . If the eigenvalues of  $J = f'(y^*)$  all lie strictly in the left complex half-plane, then the equilibrium point  $y^*$  is asymptotically stable. If J has any e-val in the right complex half plane then  $y^*$  is an unstable point.

**Dahlquist Test Equation:**  $y' = \lambda y, \quad \lambda \in \mathbb{C}$ 

- Has complex valued solution  $y(t) = e^{\lambda t} y_0$
- Equilibrium point  $y^* = 0$
- Asymptotically stable if  $\operatorname{Re}(\lambda) < 0$

**Stability Function:**  $y_{n+1} = R(h\lambda)y_n$  $R(\mu) = 1 + \mu b^T (I - \mu A)^{-1} \mathbf{1}$ 

- Euler:  $R(h\lambda) = 1 + h\lambda$ The fixed point  $y^* = 0$  is asymptotically stable if when we start near zero, we tend to it, i.e  $|1 + h\lambda| < 1$
- Implicit Euler:  $R(h\lambda) = (1 h\lambda)^{-1}$
- **Trapezium:**  $R(h\lambda) = \frac{1+h\lambda/2}{1-h\lambda/2}$
- Implicit Midpoint:  $R(h\lambda) = \frac{1+h\lambda/2}{1-h\lambda/2}$

Determine  $h\lambda$  such that  $|R(h\lambda)| \leq 1$ , this is called **the region** of absolute stability of the numerical method.

#### **Fixed Points**

Set of Fixed Points:  $\mathcal{F} = \{y \in \mathbb{R}^d : f(y) = 0\}$ For a numerical map  $\Psi_h$ , the fixed point may depend on h as well as f. We denote the set of fixed points  $\Psi_h$  by  $\mathcal{F}_h = \{y \in \mathbb{R}^d : \Psi_h(y) = y\}$ 

**A-stable:** If stability region includes entire left half-plane **L-stable:** If A-stable and  $R(\mu) \to 0$  as  $\mu \to \infty$ 

**Thm:** Given an RK method with stability function R, then the method is A-stable iff:

- All poles of R lie strictly in the right half plane
- $|R(it)| \leq 1$  for all  $t \in \mathbb{R}$

# Linear Multistep Methods

**Definition 5.0.1:** A linear k-step method is defined as:

$$\sum_{j=0}^{k} \alpha_j y_{n+j} = h \sum_{j=0}^{k} \beta_j f\left(y_{n+j}\right)$$

Where  $\alpha_k \neq 0$  and either  $\alpha_0 \neq 0$  or  $\beta_0 \neq 0$ . Usually the coefficients are normalized such that either  $\alpha_k = 1$  or  $\sum_j \beta_j = 1$ . Order of Accuracy:

$$\rho(\zeta) = \sum_{j=0}^{k} \alpha_j \zeta^j, \quad \sigma(\zeta) = \sum_{j=0}^{k} \beta_j \zeta^j$$

The  $\theta$ -method: generalises all linear one-step methods:

$$y_{n+1} - y_n = h(1 - \theta)f(y_n) + h\theta f(y_{n+1})$$

**Residual:** The residual of a linear multistep method at time  $t_{n+k}$ :

$$r_{n} := \sum_{j=0}^{k} \alpha_{j} y(t_{n+j}) - h \sum_{j=0}^{k} \beta_{j} y'(t_{n+j})$$

**Consistency:** Equivalent conditions for a linear multistep method to have order of consistency p are:

• The coefficients  $\alpha_j$  and  $\beta_j$  satisfy

$$\sum_{j=0}^{k} \alpha_{j} = 0 \text{ and } \sum_{j=0}^{k} \alpha_{j} j^{i} = i \sum_{j=0}^{k} \beta_{j} j^{i-1}$$

for  $i = 1, \ldots, p$ 

• The polynomials  $\rho(\zeta)$  and  $\sigma(\zeta)$  satisfy:

$$\rho\left(e^{z}\right) - z\sigma\left(e^{z}\right) = \mathcal{O}\left(z^{p+1}\right)$$

• The polynomials  $\rho(\zeta)$  and  $\sigma(\zeta)$  satisfy:

$$\frac{\rho(z)}{\log z} - \sigma(z) = \mathcal{O}\left((z-1)^p\right)$$

#### Root Condition and Zero-Stability

**Root Condition:** A linear k-step method is said to satisfy the root condition if all roots  $\zeta$  of  $\rho(\zeta) = 0$  lie in the unit disc  $(|\zeta| \leq 1)$  and any root of modulus one has multiplicity one. **Theorem 5.4.1:** A multistep method is convergent iff the order is  $p \geq 1$  and satisfies the root condition.

## Geometric Integration

# **Definitions**

• Abscissa: The distance from a point to the vertical or y -axis, measured parallel to the horizontal or x -axis; the x -coordinate.

- Implicit: The method is defined implicitly by an equation that has to be solved to advance the step (e.g. contains a function that depends on  $y_{n+1}$ .
- **Explicit:** Calculates the state of a system at a later time from a the state of the system at a current time.
- Consistency: A method is consistent of order p if, in a single timestep, the difference between the exact and approximate solutions is  $O(h^{p+1})$ .
- **Stability:** A method is stable if the difference between numerical solutions grows by a bounded amount as *h* tends to zero.
- Local Error: error introduced in one step of a numerical method.
- Global Error:  $ge = |y_n y(t_n)|$ .

$$e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$$

- **First Order Method:** When Euler's method is applied with fixed stepsize on a finite time interval, the norm of the global error is bounded by a constant times the stepsize.
- Local error for Euler's method satisfies:  $le_n \leq |y''(t_n)|h^2/2$
- Liouvilles Theorem: Flow maps of divergence free systems are volume preserving.